



# Reputations and Games

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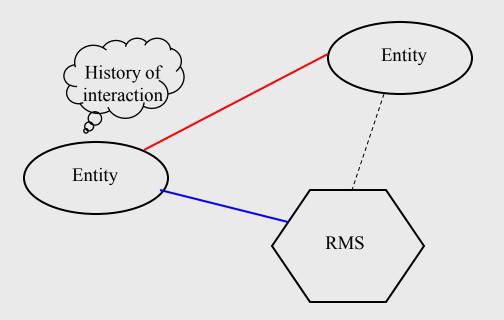
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## Reputation Management



- Applicable to P2P
- "Degree" of Trust
- Evidence-based updates
- Amenable to Decision-Theoretic Framework

Although quantitative, these approaches are *ad hoc*, informal, and provide only simplistic capabilities.





# Metrics – Challenge for security

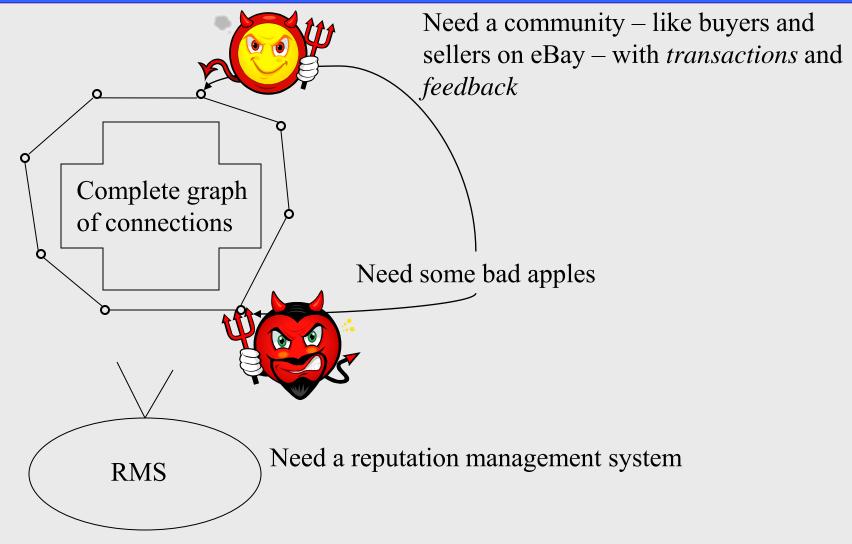
Metrics are hard for security, privacy, etc.

- Can we do better with reputation systems?
  - Would like measure of "value added" by RMS
  - Would allow us to compare RMS"s.

We decided to try this for a simple framework



### Simplest RMS framework?







### What transactions?

Simplest: "Abstract Client-Server Interactions"



- Client reports feedback to RMS.
- Bad guys could "frequently" provide bad service or bad feedback or both



### Our transactions

- File sharing
  - More complex than "abstract transactions" because of side effect of file copy
  - Practically relevant
- Initially several copies of each file; probabilistically some are corrupted
- Good users
  - "clean up" i.e., remove bad files upon receipt with high probability
  - Provide honest feedback





### Bad users

- Infinite variation possible in bad behavior
- To keep things simple we model bad users by two probabilities
  - Probability of clean-up  $\alpha$
  - Probability of honest feedback β
- What this doesn't allow:
  - Collusions
  - Targeted bad behavior



### **User Models**

- A two-dimensional approach to behavior:
  - Cleanup (%):
     Upon reception of an invalid file, how likely is user to remove that file?
  - Honesty (%):
     With what
     probability will a
     user provide
     honest feedback

User Type	Cleanup	Honesty	Source
Good	90%-100%	100%	BEST
Purely Malicious	0%-10%	0%	WORST
Feedback Malicious	90%-100%	0%	RAND
Malicious Provider	0-10%	100%	WORST
Disguised Malicious	50%-100%	50%-100%	RAND
Sybil	Sybil users particip then create a	WORST	

Note: "Source" dictates how Reputation values are used to choose a file-sender



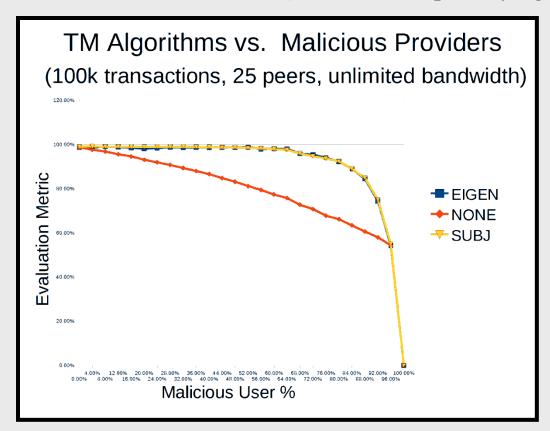
## TM systems analyzed

- None: The absence of TM, used for control runs
- EigenTrust by Hector Garcia-Molina et. al.
  - Globally convergent Reputation via matrix multiplication of normalized values
  - Convergence quick due to certain matrix properties
- Subjective Logic by Audun Jøsang et. al.
  - Triples of the form (belief, disbelief, uncertainty)
  - Transitive paths examined using "discount" and "consensus" logic operators
  - Reputation values correlate with beta-PDF functions



### Metric & Results

• Metric: (# trans. with "good" recipients, resulting in trade of valid file)
(# trans. attempted by "good" recepients)



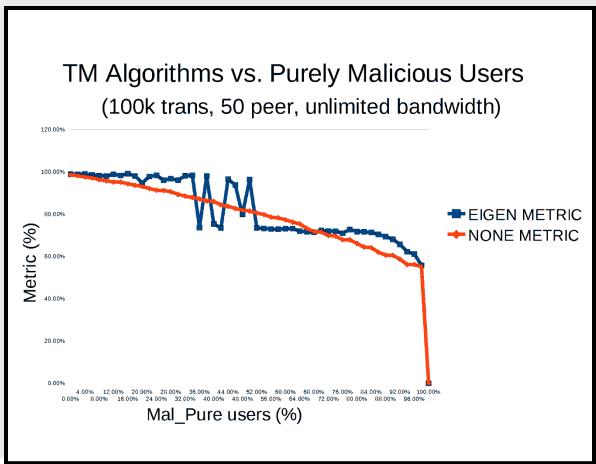
TM works well



### A More Complex Example

More interesting is when users are bad, lie about their behavior, and have other bad peers lie on their behalf (as at right).

EigenTrust in particular demonstrates some very interesting properties under varying number of bad users (a topic currently under study).







# Future Work (1)

- Adversary (BOTMaster) recruits nodes dynamically
  - Limited number of bad nodes
  - Dynamic trustworthiness how incorruptible is a node?
- Collusions/Targeted attacks
  - Who do the bad nodes give bad feedback to? Bad files to?
- Correlations
  - Do corrupting nodes share bad files with corrupted nodes?







### **CS** and Economics

## Network of Agents

#### **CS View**

- Repeated interactions between sets of players
- Good players follow protocols
- Bad players adversarial: seek to inflict harm while escaping detection
- Goal: Prevent harm; maximize system utility

#### **Econ View**

- Repeated games with strategy-based payoffs
- All players "selfinterested"
- What do "good" and "bad" mean?
- Goal: Maximize social welfare (at least welfare of "good" players)





### Payoffs in Repeated Games (Economics)

Player i plays game at each time instant. At time t receives payoff  $f_i(t)$ .

Discounted payoff to player i: (1-d)  $\sum d^t f_i(t)$ 

(Also possible to consider average payoff per game if limit exists... but discounted makes more sense usually.)



# Repeated vs One-Shot Game: Example

#### Prisoner"s Dilemma

	Confess	Silent
Confess	-6	-1
Silent	-9	-2

Payoffs to row player; Symmetrically to column player

#### **One-Shot Nash Equilibrium**

**Both Confess** 

#### Repeated Game Nash Equilibria

1) Grim strategy: Silent until opponent reneges

Confess thereafter

(Equilibrium if d high enough)

2) Both confess always





### Good vs Bad in Econ View

- Games have multiple equilibria bad players drive towards equilibria with low social welfare?
- Bad players have a small value of; they don"t care about future payoffs, and hence about "reputations"?
- alpha-altruism (New concept.) Player wants to optimize her own payoff + alpha(payoff to other players). (½-altruistic player payoff for (C,C) = -9)
- Bad Players may have evolutionary strategies to achieve objectives and mask behavior



## Reputation Manager Outline

- Monitors interactions/games by one of:
  - Observing strategies played by all players
  - Observing payoffs to each player
  - Receiving feedback from players about strategies/payoffs
- Player strategies are function of advise from Reputation Manager and past history



# Issues for Reputation Manager

- Thwart adversaries; enhance experience for good players
- Identify players who are bad in the senses defined above and "warn" good players
- Prevent following problems
  - Whitewash: acquire new identity after bad behavior
  - Phantom feedback: acquire multiple identities and provide spurious feedback to skew reputations
  - No feedback: fail to provide feedback when due



# Evaluating Reputation Managers

#### CS View:

- Prevent harmful attacks
- Make system available/useful to good players

#### Econ View:

- Increase total welfare to good players
- Possibly enforce fair sharing of welfare among good players?



# Future Work (2)

- Altruistic Players: An exciting new model? Explore
- Can reputation managers identify the altruism parameter of each player in an arbitrary game?
   Based on what observations?
- What if a player"s degree of altruism is altered by nature of opponent?
- Analyze games under other notions of "badness"
- Reconcile Econ view to real systems?? Where do we get payoffs, lists of strategies from?

